

## ✓ K means clustering

```
import pandas as pd
df = pd.read_csv('/kaggle/input/unsupervised-learning-on-country-data/Country-data.csv')
print(df.head())
```

```

country  child_mort  exports  health  imports  income \
0  Afghanistan    90.2    10.0    7.58    44.9    1610
1    Albania      16.6    28.0    6.55    48.6    9930
2    Algeria      27.3    38.4    4.17    31.4   12900
3    Angola      119.0    62.3    2.85    42.9    5900
4  Antigua and Barbuda  10.3    45.5    6.03    58.9   19100

inflation  life_expec  total_fer  gdpp
0      9.44      56.2      5.82   553
1      4.49      76.3      1.65  4090
2     16.10      76.5      2.89  4460
3     22.40      60.1      6.16  3530
4      1.44      76.8      2.13 12200
```

```
from sklearn.preprocessing import OneHotEncoder
```

```
ohe = OneHotEncoder(sparse=False)
```

```
country_encoded = ohe.fit_transform(df[['country']])
```

```

/opt/conda/lib/python3.10/site-packages/sklearn/preprocessing/_encoders.py:868: FutureWarning:
warnings.warn(
```

```
country_df = pd.DataFrame(country_encoded, columns=ohe.get_feature_names_out(['country']))
```

```
print(country_df.head())
```

```

country_Afghanistan  country_Albania  country_Algeria  country_Angola \
0                1.0                0.0                0.0                0.0
1                0.0                1.0                0.0                0.0
2                0.0                0.0                1.0                0.0
3                0.0                0.0                0.0                1.0
4                0.0                0.0                0.0                0.0

country_Antigua and Barbuda  country_Argentina  country_Armenia \
0                0.0                0.0                0.0
1                0.0                0.0                0.0
2                0.0                0.0                0.0
3                0.0                0.0                0.0
4                1.0                0.0                0.0

country_Australia  country_Austria  country_Azerbaijan  ... \
0                0.0                0.0                0.0  ...
1                0.0                0.0                0.0  ...
2                0.0                0.0                0.0  ...
3                0.0                0.0                0.0  ...
4                0.0                0.0                0.0  ...
```

```

country_United Arab Emirates  country_United Kingdom  \
0                             0.0                      0.0
1                             0.0                      0.0
2                             0.0                      0.0
3                             0.0                      0.0
4                             0.0                      0.0

country_United States  country_Uruguay  country_Uzbekistan  \
0                      0.0             0.0                   0.0
1                      0.0             0.0                   0.0
2                      0.0             0.0                   0.0
3                      0.0             0.0                   0.0
4                      0.0             0.0                   0.0

country_Vanuatu  country_Venezuela  country_Vietnam  country_Yemen  \
0              0.0                 0.0                0.0        0.0
1              0.0                 0.0                0.0        0.0
2              0.0                 0.0                0.0        0.0
3              0.0                 0.0                0.0        0.0
4              0.0                 0.0                0.0        0.0

country_Zambia
0              0.0
1              0.0
2              0.0
3              0.0
4              0.0

[5 rows x 167 columns]

```

```
print(country_df.info())
```

```

➡ <class 'pandas.core.frame.DataFrame'>
RangeIndex: 167 entries, 0 to 166
Columns: 167 entries, country_Afghanistan to country_Zambia
dtypes: float64(167)
memory usage: 218.0 KB
None

```

```
df = pd.concat([df.drop('country',axis=1),country_df],axis=1)
```

```
print(df.head())
```

```

➡
  child_mort  exports  health  imports  income  inflation  life_expec  \
0         90.2    10.0    7.58    44.9    1610         9.44    56.2
1         16.6    28.0    6.55    48.6    9930         4.49    76.3
2         27.3    38.4    4.17    31.4   12900        16.10    76.5
3        119.0    62.3    2.85    42.9    5900        22.40    60.1
4         10.3    45.5    6.03    58.9   19100         1.44    76.8

  total_fer  gdpp  country_Afghanistan  ...  country_United Arab Emirates  \
0         5.82   553                  1.0  ...                          0.0
1         1.65  4090                  0.0  ...                          0.0
2         2.89  4460                  0.0  ...                          0.0
3         6.16  3530                  0.0  ...                          0.0
4         2.13 12200                  0.0  ...                          0.0

  country_United Kingdom  country_United States  country_Uruguay  \
0                      0.0                    0.0                0.0
1                      0.0                    0.0                0.0

```

2	0.0	0.0	0.0
3	0.0	0.0	0.0
4	0.0	0.0	0.0

	country_Uzbekistan	country_Vanuatu	country_Venezuela	country_Vietnam	\
0	0.0	0.0	0.0	0.0	
1	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	

	country_Yemen	country_Zambia
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0

[5 rows x 176 columns]

```
print(df.info())
```

```
>>> <class 'pandas.core.frame.DataFrame'>
RangeIndex: 167 entries, 0 to 166
Columns: 176 entries, child_mort to country_Zambia
dtypes: float64(174), int64(2)
memory usage: 229.8 KB
None
```

```
print(df.shape)
```

```
>>> (167, 176)
```

```
from sklearn.cluster import KMeans
```

```
wcss = []
for i in range(1,11):
    kmeans = KMeans(n_clusters=i, random_state=42)
    kmeans.fit(df)
    wcss.append(kmeans.inertia_)
```

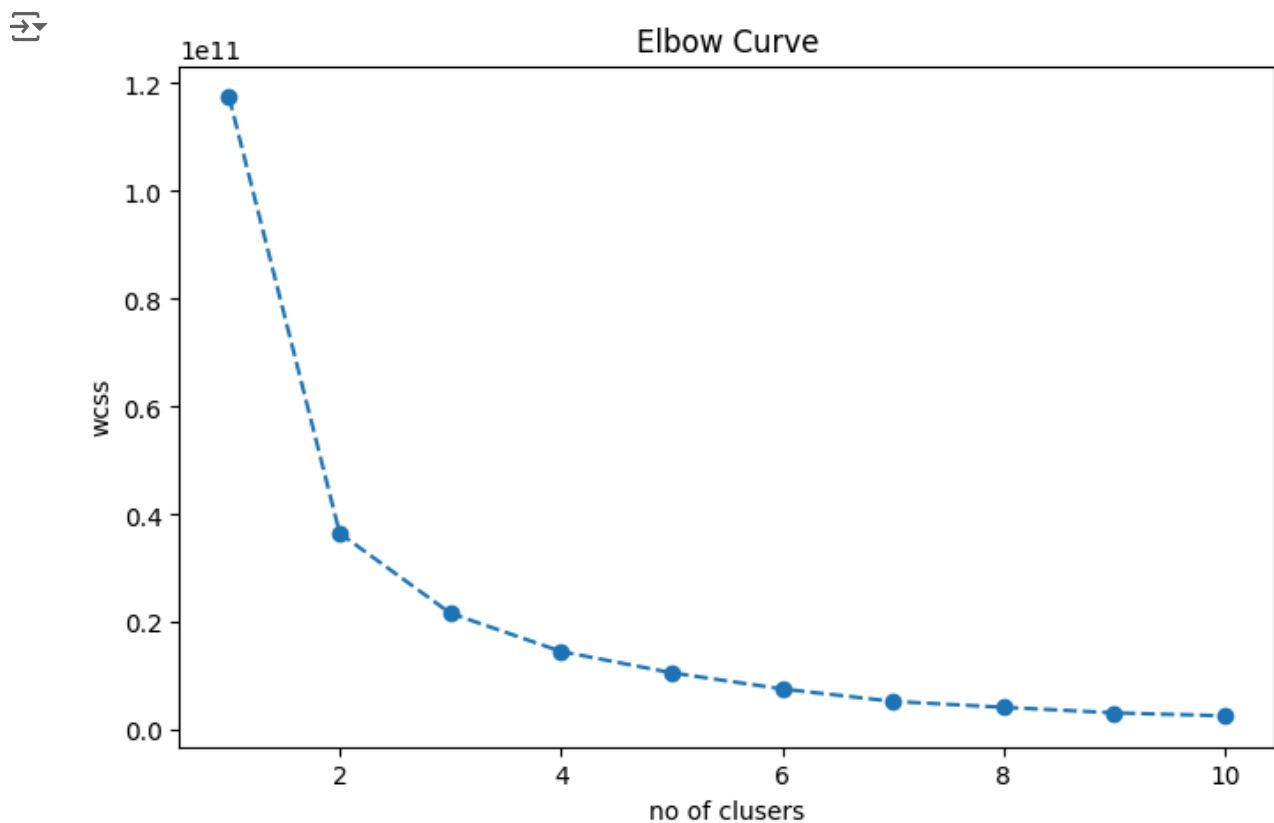
```
>>> /opt/conda/lib/python3.10/site-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The c
warnings.warn(
/opt/conda/lib/python3.10/site-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The c
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warnings.warn(
```

/opt/conda/lib/python3.10/site-packages/sklearn/cluster/\_kmeans.py:870: FutureWarning: The c  
warnings.warn()

```
print(wcss)
```

```
[117459687635.07814, 36528388099.3221, 21710210203.75414, 14610713609.535135, 10655378245.86]
```

```
#plot elbow curve
import matplotlib.pyplot as plt
plt.figure(figsize=(8,5))
plt.plot(range(1,11),wcss,marker='o',linestyle='--')
plt.title("Elbow Curve")
plt.xlabel("no of clusers")
plt.ylabel('wcss')
plt.show()
```



```
kmeans = KMeans(n_clusters=4,random_state=42)
y_kmeans = kmeans.fit_predict(df)
df['cluster']=y_kmeans
```

```
/opt/conda/lib/python3.10/site-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The c  
warnings.warn()
```

```
print(df.head())
```

```
child_mort  exports  health  imports  income  inflation  life_expec  \
0          90.2    10.0    7.58    44.9    1610         9.44    56.2
1          16.6    28.0    6.55    48.6    9930         4.49    76.3
```

2	27.3	38.4	4.17	31.4	12900	16.10	76.5
3	119.0	62.3	2.85	42.9	5900	22.40	60.1
4	10.3	45.5	6.03	58.9	19100	1.44	76.8

	total_fer	gdpp	country_Afghanistan	...	country_United Kingdom	\
0	5.82	553	1.0	...		0.0
1	1.65	4090	0.0	...		0.0
2	2.89	4460	0.0	...		0.0
3	6.16	3530	0.0	...		0.0
4	2.13	12200	0.0	...		0.0

	country_United States	country_Uruguay	country_Uzbekistan	\
0	0.0	0.0	0.0	
1	0.0	0.0	0.0	
2	0.0	0.0	0.0	
3	0.0	0.0	0.0	
4	0.0	0.0	0.0	

	country_Vanuatu	country_Venezuela	country_Vietnam	country_Yemen	\
0	0.0	0.0	0.0	0.0	
1	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	

	country_Zambia	cluster
0	0.0	0
1	0.0	0
2	0.0	0
3	0.0	0
4	0.0	2

[5 rows x 177 columns]

```
labels = kmeans.labels_
```


```
from sklearn.metrics import silhouette_score
```

```
sil_score = silhouette_score(df, labels)
print(f'Silhouette Score: {sil_score}')
```

 Silhouette Score: 0.6109297327684861

## ✓ Agglomerative Clustering

```
print(df.head())
```



0	90.2	10.0	7.58	44.9	1610	9.44	56.2
1	16.6	28.0	6.55	48.6	9930	4.49	76.3
2	27.3	38.4	4.17	31.4	12900	16.10	76.5
3	119.0	62.3	2.85	42.9	5900	22.40	60.1
4	10.3	45.5	6.03	58.9	19100	1.44	76.8

	total_fer	gdpp	country_Afghanistan	...	country_United Kingdom	\
0	5.82	553	1.0	...		0.0
1	1.65	4090	0.0	...		0.0
2	2.89	4460	0.0	...		0.0

```

3      6.16   3530      0.0   ...      0.0
4      2.13  12200      0.0   ...      0.0

```

```

country_United States country_Uruguay country_Uzbekistan \
0      0.0      0.0      0.0
1      0.0      0.0      0.0
2      0.0      0.0      0.0
3      0.0      0.0      0.0
4      0.0      0.0      0.0

```

```

country_Vanuatu country_Venezuela country_Vietnam country_Yemen \
0      0.0      0.0      0.0      0.0
1      0.0      0.0      0.0      0.0
2      0.0      0.0      0.0      0.0
3      0.0      0.0      0.0      0.0
4      0.0      0.0      0.0      0.0

```

```

country_Zambia cluster
0      0.0      0
1      0.0      0
2      0.0      0
3      0.0      0
4      0.0      2

```

[5 rows x 177 columns]

```
print(df.shape)
```

```
(167, 177)
```

```
from sklearn.cluster import AgglomerativeClustering
```

```
ac = AgglomerativeClustering(n_clusters=3,affinity='euclidean',linkage='ward')
```

```
labels = ac.fit_predict(df)
```

```
/opt/conda/lib/python3.10/site-packages/sklearn/cluster/_agglomerative.py:983: FutureWarning:
warnings.warn(
```

```
print(labels)
```

```

[2 2 2 2 1 1 2 0 0 2 1 1 2 1 2 0 2 2 2 2 2 2 1 0 2 2 2 2 2 0 2 2 2 1 2 2 2
 2 2 2 2 1 1 1 0 2 2 2 2 1 2 1 2 0 0 2 2 2 0 2 1 2 2 2 2 2 2 1 0 2 2 2 2 0
 1 0 2 0 2 1 2 2 0 2 2 1 2 2 2 1 1 0 2 2 2 1 2 2 1 2 2 2 2 2 2 2 2 2 2 2 0
 1 2 2 0 1 2 2 2 2 2 1 1 0 2 1 2 2 1 2 2 1 2 0 1 1 2 2 1 1 2 2 2 2 0 0 2 2
 2 2 2 2 2 1 2 2 2 0 0 0 1 2 2 1 2 2 2]

```

```
from sklearn.metrics import silhouette_score
```

```
print(f"silhouette_score: {silhouette_score(df,labels)}")
```

```
silhouette_score: 0.5963166843614005
```

